

TUTORIAL

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On Your Seats (1/4)

Write Correct / Incorrect, Tight / Not Tight

- | | |
|---|-----------------------------|
| 1. $4n^2 - 300n + 12 \in O(n^2)$ | CORRECT, TIGHT |
| 2. $4n^2 - 300n + 12 \in O(n^3)$ | CORRECT, NOT TIGHT |
| 3. $3^n + 5n^2 - 3n \in O(n^2)$ | INCORRECT, NOT TIGHT |
| 4. $3^n + 5n^2 - 3n \in O(4^n)$ | CORRECT, NOT TIGHT |
| 5. $3^n + 5n^2 - 3n \in O(3^n)$ | CORRECT, TIGHT |
| 6. $50 \cdot 2^n n^2 + 5n - \log(n) \in O(2^n)$ | INCORRECT, NOT TIGHT |



Exercise 2

Write True or False :

$$T(n) = 5n^3 + 2n^2 + 4 \log n$$

1. $T(n) \in O(n^4)$
2. $T(n) \in O(n^2)$
3. $T(n) \in \Theta(n^3)$
4. $T(n) \in O(\log n)$
5. $T(n) \in \Theta(n^4)$
6. $T(n) \in \Omega(n^2)$

- Rules of thumb
- in describing the asymptotic complexity of an algorithm:
 - If the running time is the sum of multiple terms, keep the one with the largest growth rate and drop the others, since they will not have an impact for large
 - If the remaining term is a product, drop any multiplicative constants

$$f(n) \approx 1000 + n + 2n^2 \mapsto f(n) \approx n^2$$

$$f(n) \approx 2^n + 2^{n+1} \mapsto f(n) \approx 2^{n+1}$$

$$f(n) \approx 2 + 400n + 2^n \mapsto f(n) \approx 2^n$$

$$f(n) \approx n + 500 \log n \mapsto f(n) \approx n$$

$$f(n) \approx 2\sqrt{n} + 500 \log n \mapsto f(n) \approx \sqrt{n}$$

$$f(n) \approx \sqrt{n} + n + n^3 \mapsto f(n) \approx n^3$$

MATH BACKGROUND: EXPONENTS

■ Some useful identities:

- $X^A \cdot X^B = X^{A+B}$
- $X^A / X^B = X^{A-B}$
- $(X^A)^B = X^{AB}$
- $X^N + X^N = 2X^N$
- $2^N + 2^N = 2^{N+1}$

MATH BACKGROUND: LOGARITHMS

■ Logarithms

- *definition:* $X^A = B$ if and only if $\log_X B = A$
- *intuition:* $\log_X B$ means:
"the power X must be raised to, to get B "
- In this course, a logarithm with no base implies base 2.
 $\log B$ means $\log_2 B$

■ Examples

- $\log_2 16 = 4$ (because $2^4 = 16$)
- $\log_{10} 1000 = 3$ (because $10^3 = 1000$)

- **$O(1)$** : Time complexity of a function (or set of statements) is considered as $O(1)$ if it doesn't contain loop, recursion, and call to any other non-constant time function.
- **$O(n)$** : Time Complexity of a loop is considered as $O(n)$ if the loop variables are incremented/decremented by a constant amount. For example following functions have $O(n)$ time complexity.

```
// Here c is a positive integer constant
for (int i = 1; i <= n; i += c) {
    // some  $O(1)$  expressions
}
```

- **$O(n^2)$** : Time complexity of nested loops is equal to the number of times the innermost statement is executed. For example, the following sample loops have $O(n^2)$ time complexity

```
for (int i = 1; i <= n; i += c) {
    for (int j = 1; j <= n; j += c) {
        // some  $O(1)$  expressions
    }
}

for (int i = n; i > 0; i -= c) {
    for (int j = i+1; j <= n; j += c) {
        // some  $O(1)$  expressions
    }
}
```

- **O(Logn)** Time Complexity of a loop is considered as $O(\text{Log}n)$ if the loop variables are divided/multiplied by a constant amount.

```
for (int i = 1; i <= n; i *= c) {
    // some O(1) expressions
}
for (int i = n; i > 0; i /= c) {
    // some O(1) expressions
}
```

- What is the exact runtime and complexity class (Big-Oh)?

```
int sum = 0;
for (int i = 1; i <= N; i += c) {
    sum++;
}
```

- Runtime = $N / c = O(N)$.

```
int sum = 0;
for (int i = 1; i <= N; i *= c) {
    sum++;
}
```

- Runtime = $\log_c N = O(\log N)$.

Call this number of multiplications "x".

$$2^x = N$$

$$x = \log_2 N$$

- After getting the above problems. Let's have two iterators in which, outer one runs **$N/2$ times**, and we know that the time complexity of a loop is considered as **$O(\log N)$** , if the iterator is divided / multiplied by a constant amount **K** then the time complexity is considered as **$O(\log_K N)$** .
- $(N/2)^K = 1$ (for k iterations)
 $\Rightarrow N = 2^k$ (taking log on both sides)
 $\Rightarrow k = \log(N)$ base 2.
 Therefore, the time complexity will be
 $T(N) = O(\log N)$

EXAMPLE: $O(N^2)$

```

6 public static void main(String[] args) {
7     int n = 100;
8     for (int i = 1; i <= n/3; i++) {
9         for (int j = 1; j < n; j=j+4) {
10             System.out.println("*");
11             break;
12         }
13     }
14 }

```

outer loop will run $n/3$ times

inner loop will run $n/4$ times

so total time complexity is $(n/3) * (n/4) = n^2/12 = O(n^2)$

- What is the exact runtime complexity (Big-Oh)?

```
int sum = 0;
for (int i = 1; i <= N; i++) {
    for (int j = 1; j <= N * 2; j++) {
        sum++;
    }
}
```

- Runtime = $N \cdot 2N = O(N^2)$.

```
int sum = 0;
for (int i = 1; i <= N; i++) {
    for (int j = 1; j <= i; j++) {
        sum++;
    }
}
```

- Arithmetic series:

$$\sum_{k=1}^n k = 1 + 2 + \dots + n = \frac{n(n+1)}{2}$$

- Runtime = $N(N + 1) / 2 = O(N^2)$.

```
if (value % 2 == 0){
    return true;
}
else
    return false;
}
```

Answer: $O(1)$. Constant run time complexity.

Because you're only ever taking one value, there is no "loop" to go through.

```
for (let i=0; i<array.length; i++) {
    if (array[i] === item) {
        return i;
    }
}
```

Answer: $O(n)$. Linear run time complexity.

HOW TO FIND COMPLEXITY?

Some rules of thumb

Basically just count the number of statements executed:

- If there are only a small number of simple statements in a program — $\mathbf{O(1)}$
- If there is a 'for' loop dictated by a loop index that goes up to n — $\mathbf{O(n)}$
- If there is a nested 'for' loop with outer one controlled by n and the inner one controlled by m — $\mathbf{O(n*m)}$
- For a loop with a range of values n , and each iteration reduces the range by a fixed constant fraction (eg: $\frac{1}{2}$) — $\mathbf{O(\log n)}$